

Recipe for Success:

Data-Driven Strategies for New & Prospective Restaurant Owners in San Diego

By Tia Page



Overview

<u>Introduction</u>	Defining the problem & importance of topic
<u>Data Collection</u>	Resources and techniques used to obtain the
	data needed for analysis
Analysis of Data	Valuable findings and insights gathered
	through data explorations
<u>Conclusion</u>	Major takeaways from analysis that can help
	aspiring restaurant owners



Introduction

Problem: San Diego restaurant closures

Causes: Highly competitive area for restaurants, overly saturated market, high costs of living

Solution: High revenues earned as a result of attracting customers and establishing popularity

Project Goals: Identify the key factors of successful restaurants in San Diego, recommend optimal locations to establish business, and find ways to enhance customer experience

Data Collection

- Data Collected: Restaurant reviews, Demographic information, geographic details, restaurant characteristics
- Collection methods: Webscraping, Data Extraction
- Tools: BeautifulSoup, Octospare data extractor
- Websites: Google Reviews, Point2Homes.com, Bklyndesigns.com, usa.com
- Final dataset: sd_reviews



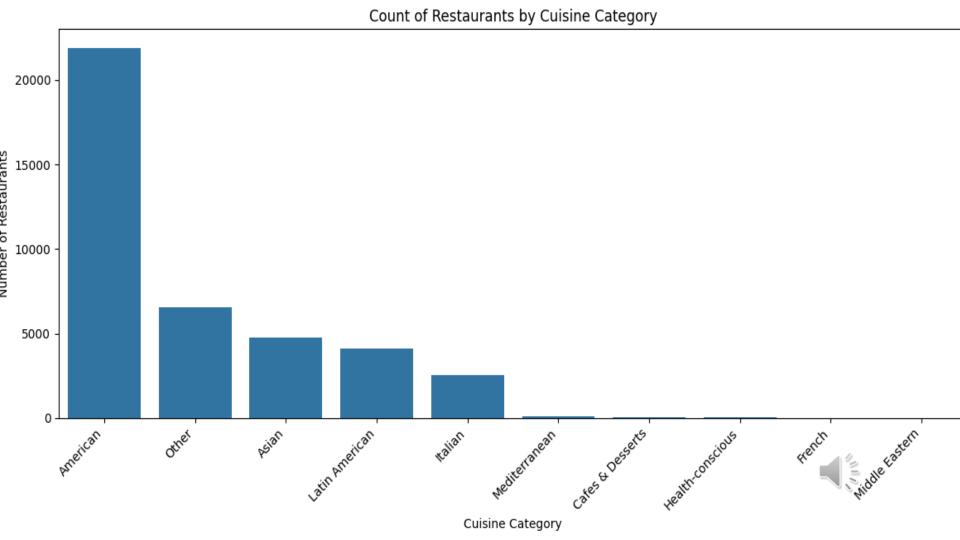
Data Cleaning & Preparation

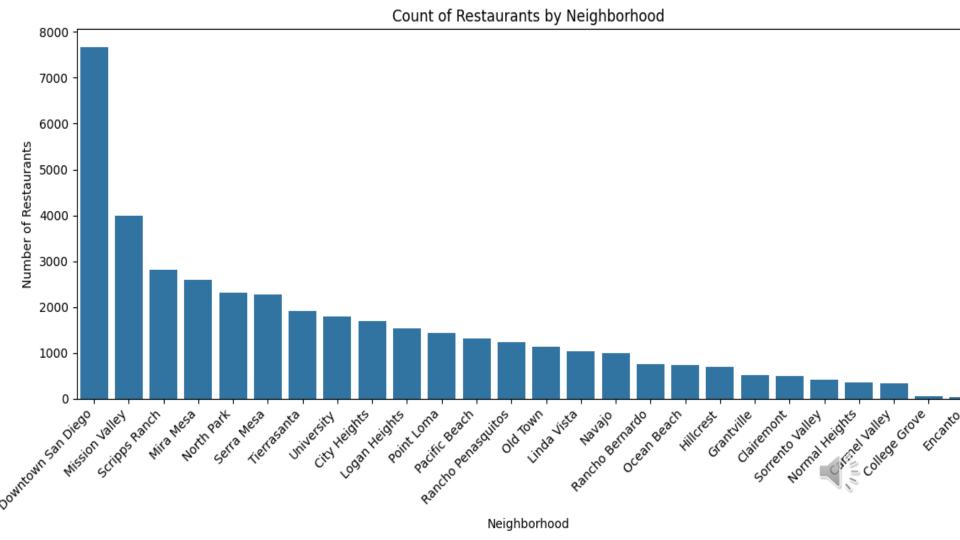
- Removed unnecessary columns: Review likes, review urls, review response
- Excluded rows with missing values
 - Some data failed to extract
- Incorrect/missing cuisine categories replaced using keywords in names
- Merged separate datasets obtained via data extraction and web scraping

EDA: Overview of Data Set

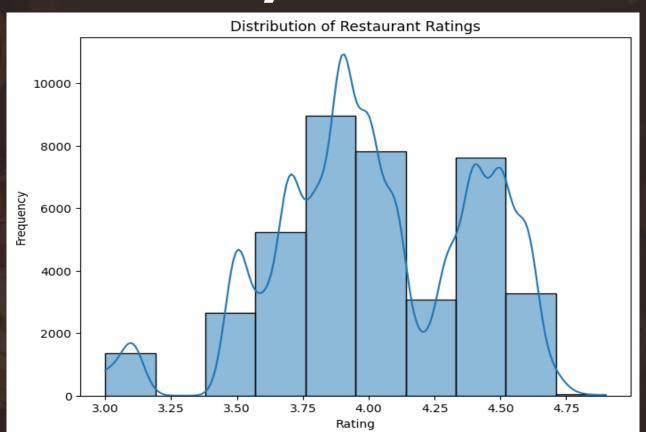
- 42,768 restaurants (rows)
- 16 variables measures (columns)
- Mixture of categorical & numeric data

```
<class 'pandas.core.frame.DataFrame'>
Index: 42768 entries, 501 to 58042
Data columns (total 16 columns):
    Column
                     Non-Null Count
                                     Dtype
                    42768 non-null
                                     object
    restaurant name
     rating
                     42768 non-null
                                     float64
    review count
                     42768 non-null
                                     object
    lowest price
                     42768 non-null
                                     float64
    highest price
                      42768 non-null
                                     float64
    cuisine type
                                     object
                     42768 non-null
    zip code
                     42768 non-null
                                     int64
    neighborhood
                      42768 non-null
                                     object
    land area
                     42768 non-null
                                     object
     population
                                     float64
                      42768 non-null
    num households
                     42768 non-null float64
    median income
                     42768 non-null float64
                      42768 non-rull float64
     average income
                     40154 non-noll object
 13
    review rating
                     40020 non-null
 14
    review
                                     object
    review likes
                      9642 non-null
                                     float64
dtypes: float64(8), int64(1), object(7)
```





Analysis of Data







#1. Identifying key factors of successful restaurants using Random Forest Classifier Model

```
Model Accuracy: 1.0
Feature Importance:
                              Feature
                                        Importance
                               rating
                                         0.861155
                         review count
                                         0.105908
               cuisine category Asian
                                         0.016447
             cuisine category Italian
                                         0.008091
10
               cuisine category Other
                                         0.004074
      cuisine category Latin American
                                         0.003255
       cuisine category Mediterranean
                                          0.000499
    cuisine category Cafes & Desserts
                                          0.000429
    cuisine category Health-conscious
                                          0.000109
      cuisine category Middle Eastern
                                          0.000018
              cuisine category French
                                          0.000015
```

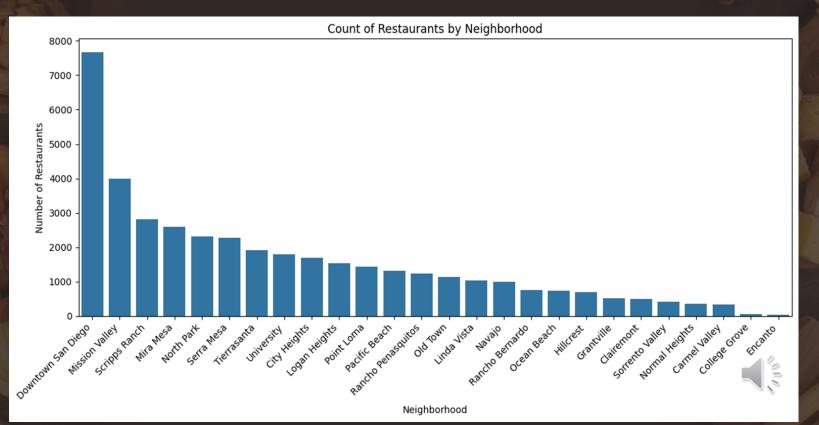


#2. Enhancing Customer Experience through TextBlob Sentiment Analysis Model

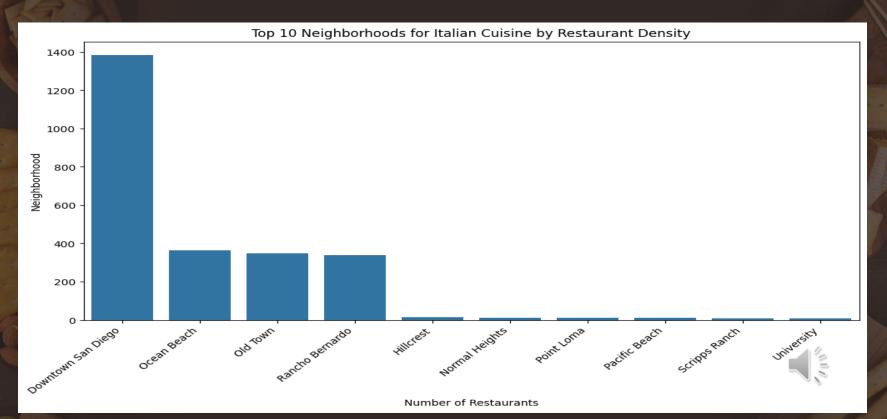




#3. Choosing an optimal restaurant location



#3. Choosing an optimal restaurant location (continued): Italian restaurant scenario



Conclusion: Main Takeaways

- Avoid competition by establishing restaurant in area with lowest density of restaurants based on specific food categories
- Enhance customer experience by first ensuring that customer service is of highest quality, then improving on food quality
- Stand out from the crowd by paying attention to the least popular food categories presented in data visualization; this also shows areas with less competition

